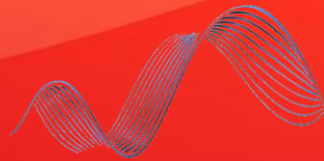
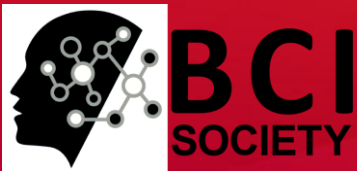


Building Foundation Models for BCIs: *Fundamentals & Opportunities*

Hatem Abou-Zeid
Assistant Professor, ESE Dept



WAVES LAB

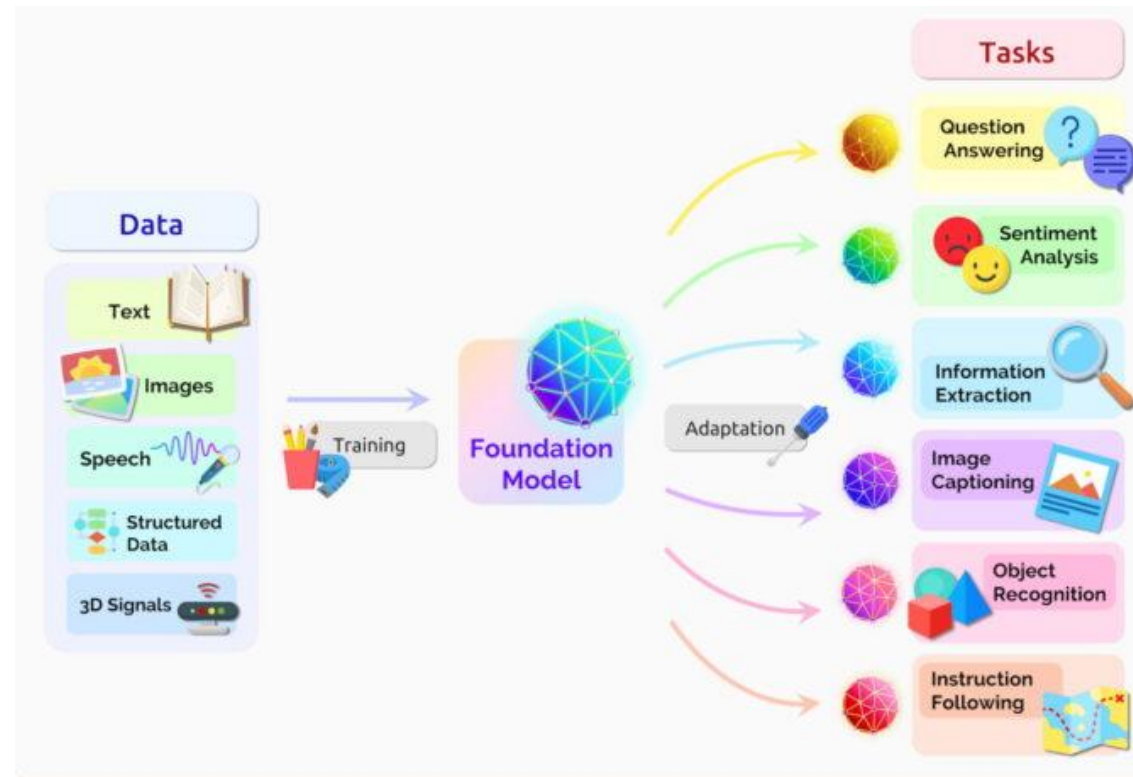


W7: Bridging AI Expertise with BCI Champions to
Advance AI Innovation in BCIs, June 2025



Foundation Models (FM) – A paradigm shift in AI

- Foundation models are (usually) large, pre-trained AI models that enable a wide range of tasks without needing lots of labeled data.
- ChatGPT is built on a foundation model



<https://blogs.nvidia.com/blog/what-are-foundation-models/>

Supervised Learning (Traditional)

1. Collect data and LABEL it (e.g. left & right motions)
2. Build an AI model for this data

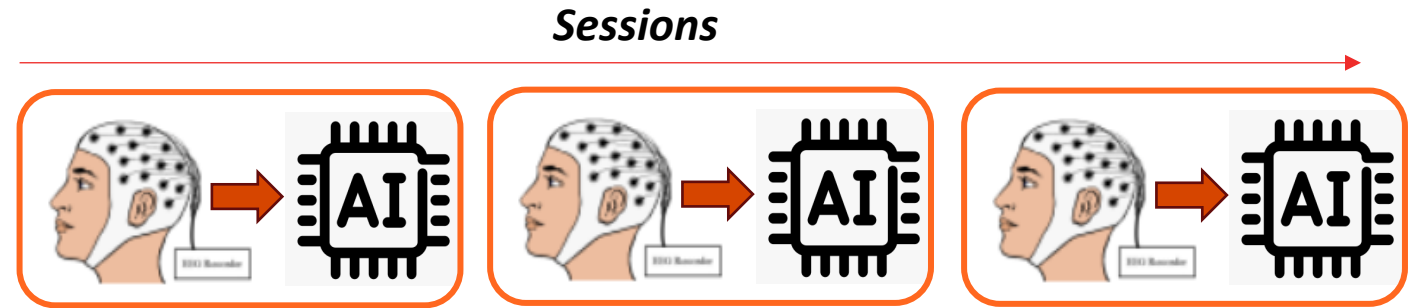


AI Generalization Challenges in BCIs

1. Models do NOT
generalize to different BCI
sessions

Considerable “calibration time”

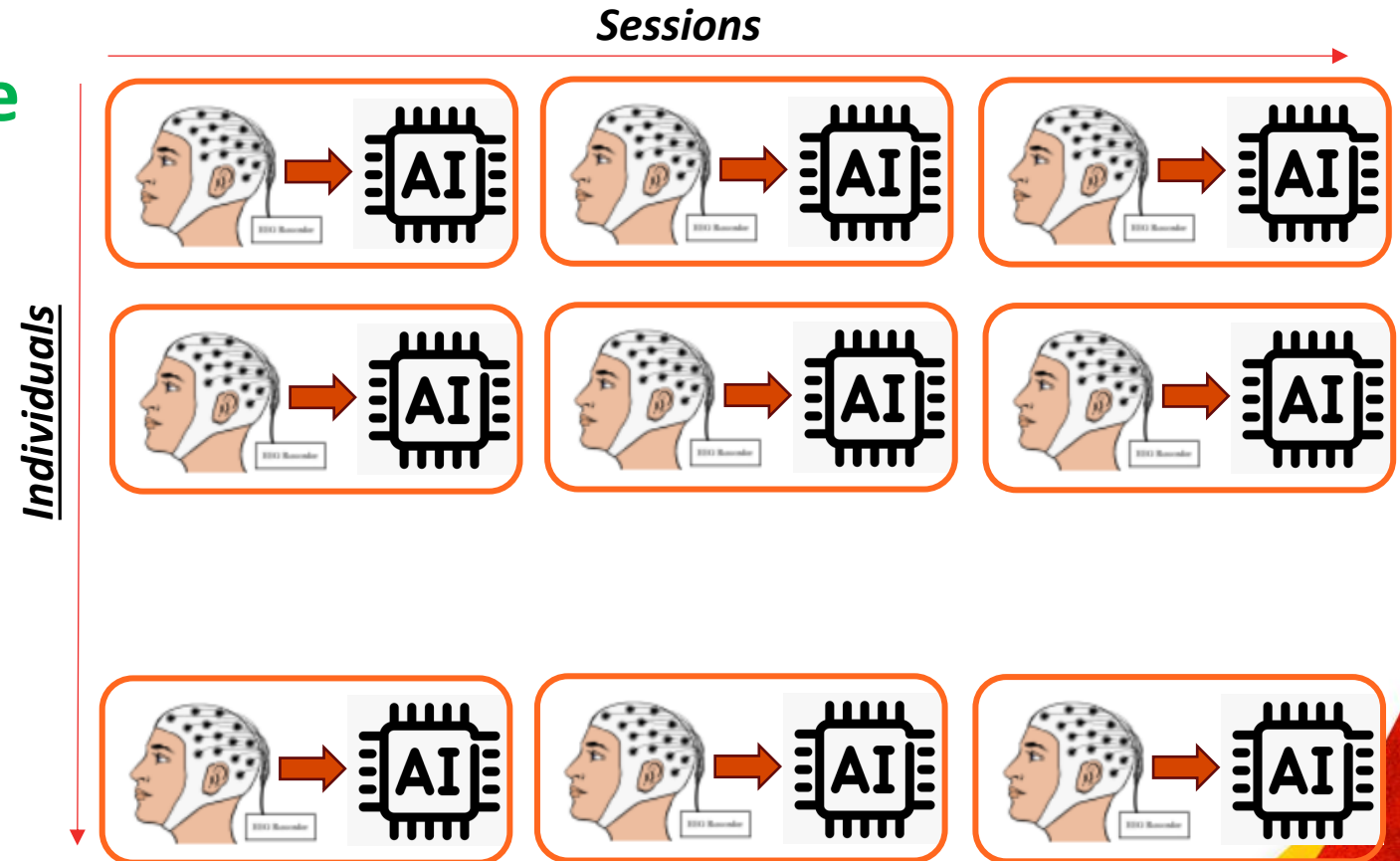
Calibration is tiring.



AI Generalization Challenges in BCIs

2. Models do not generalize to different *individuals*

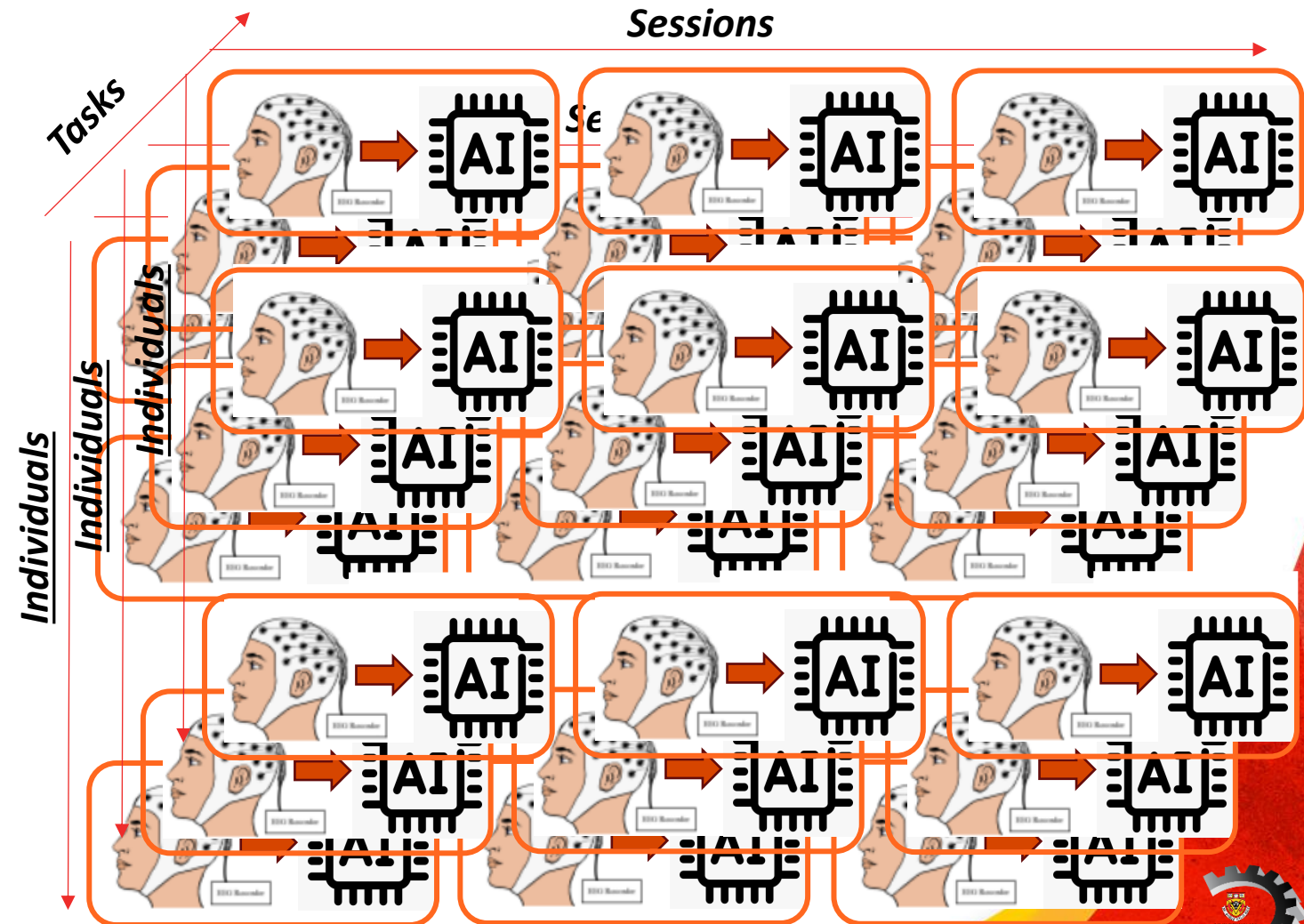
A model per individual is needed!



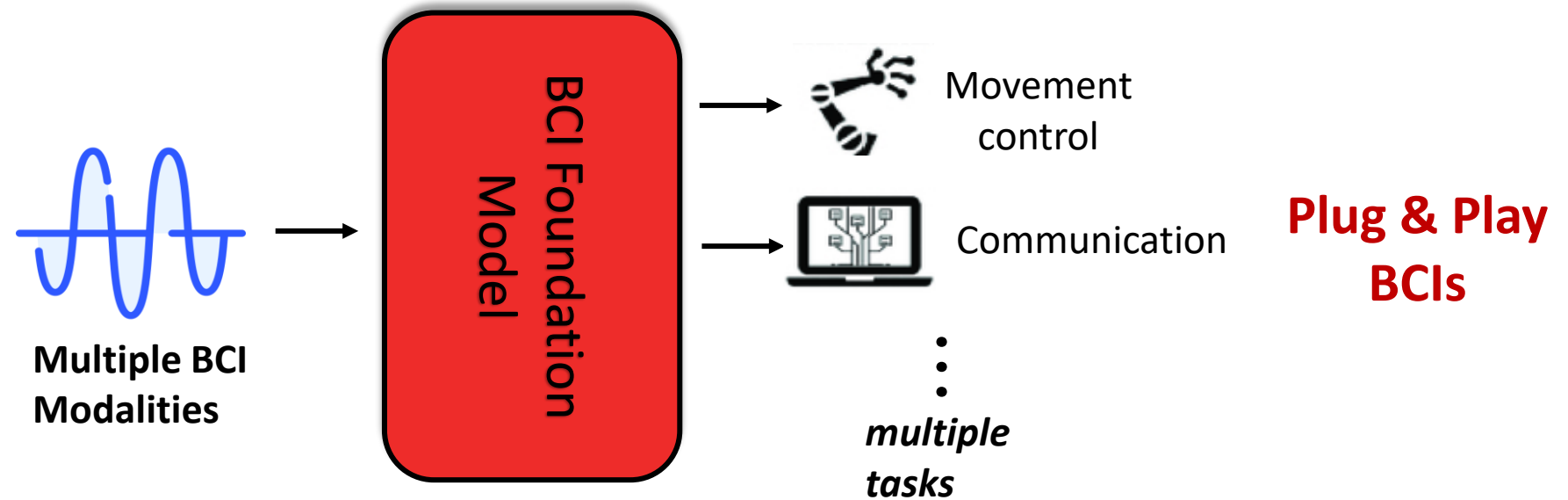
AI Generalization Challenges in BCIs

3. Models do not generalize to different tasks and modalities

A model per task per individual that is calibrated per session!

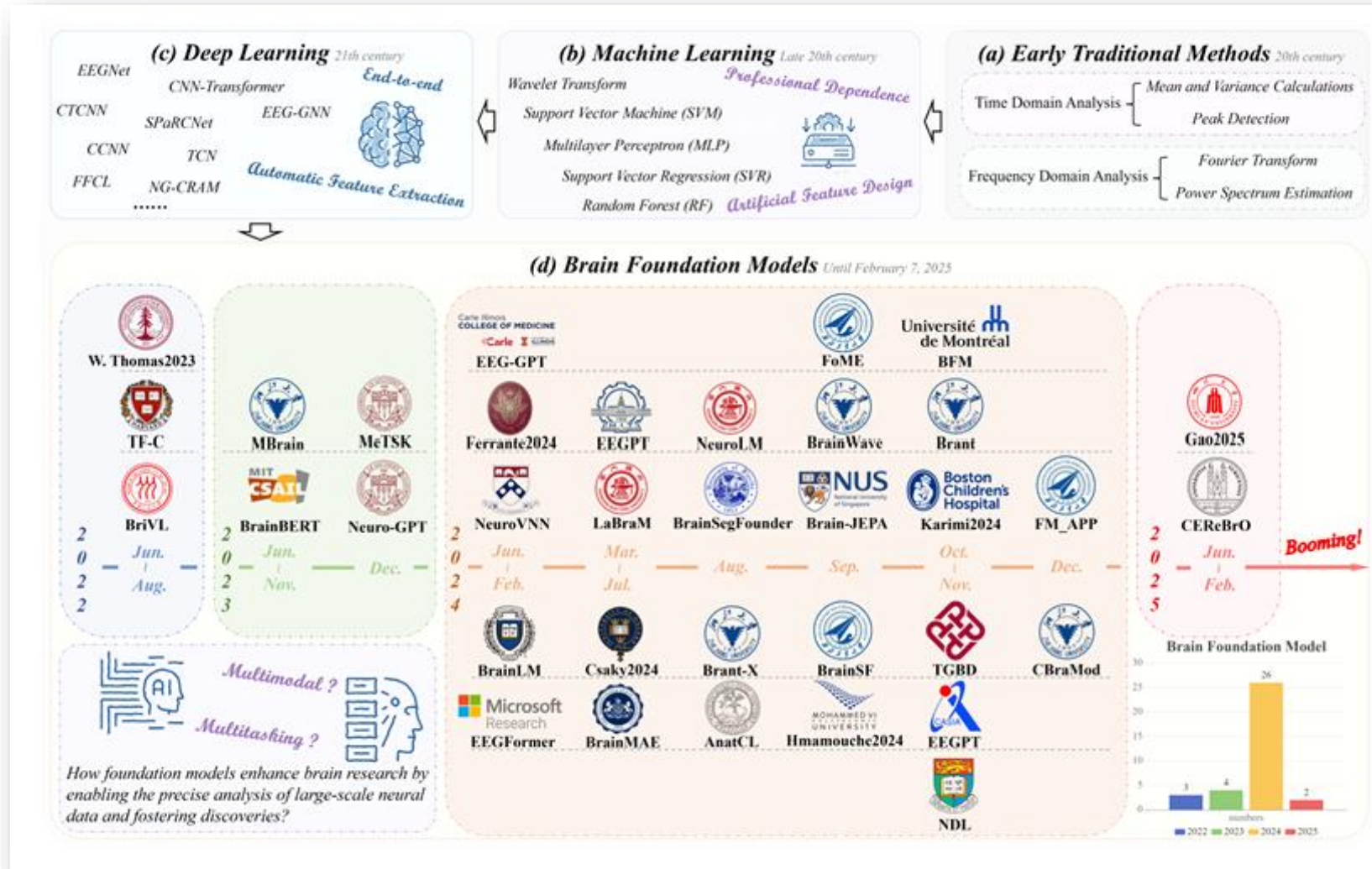


(Why) Build BCI Foundation Models



- **Generalization:** The model has learned from a **diverse range of data**, so it performs well on **unseen scenarios and new tasks**
- **Data Efficiency:** Traditional models need lots of labeled data, **foundation models perform well with small datasets.**
- **Scalability:** Instead of training a new model for each task, **one powerful model** can handle multiple tasks.

BCI Foundation Models are happening!



Zhou, Xinliang, et al. "Brain Foundation Models: A Survey on Advancements in Neural Signal Processing and Brain Discovery." *arXiv preprint arXiv:2503.00580* (2025).

BCI Foundation Models are happening!

TABLE II
SUMMARY OF TYPICAL BFMS FOR BRAIN DECODING.

Model	Pretraining	Fine-tuning	Cross Scenarios	Cross Tasks	Multi Modalities
BrainLM [19]	✓	×	✓	✓	×
LaBraM [25]	✓	✓	✓	✓	×
NeuroLM [20]	✓	×	✓	✓	✓
Brant [14]	✓	✓	✓	✓	×
BrainSegFounder [26]	✓	✓	×	×	✓
MeTSK [27]	✓	✓	×	✓	×
AnatCL [28]	✓	×	✓	✓	×
BRAINBERT [29]	✓	✓	✓	✓	×
NeuroVNN [29]	✓	✓	✓	×	×
Brain-JEPA [13]	✓	✓	×	✓	×
FoME [30]	✓	✓	✓	✓	×
FM-BIM [31]	✓	✓	×	✓	×
BrainWave [32]	✓	✓	✓	✓	✓
TF-C [33]	✓	✓	×	✓	✓
Neuro-GPT [34]	✓	✓	✓	×	×
BrainMAE [35]	✓	✓	✓	✓	×
EEGFormer [36]	✓	✓	×	✓	×
CEReBro [37]	✓	✓	×	✓	×
CBraMod [16]	✓	✓	✓	✓	×
EEGPT-Yue [38]	✓	✓	✓	✓	×
TGBD [23]	✓	×	✓	×	×
MBrain [21]	✓	✓	✓	×	✓
EEGPT [17]	✓	✓	×	✓	×
Brant-X [22]	✓	✓	×	✓	✓
FM-APP [39]	✓	×	×	✓	✓

"Cross scenarios": the ability of BFMs to operate effectively across diverse settings and conditions within a single paradigm;

"Cross tasks capability": BFMs can achieve robust performance across different paradigms.

Zhou, Xinliang, et al. "Brain Foundation Models: A Survey on Advancements in Neural Signal Processing and Brain Discovery." *arXiv preprint arXiv:2503.00580* (2025).

Rest of the talk

1. Fundamentals

- ✓ Why Foundation Models for BCI?
- Building Foundation Models (Pretraining)
- Using BCI Foundation Models (Fine-tuning)

Terminology for upcoming talks & discussions: *Autoencoders, latent representations, embeddings, SSL, contrastive learning, masked prediction, autoregression, foundation models, downstream tasks, frozen encoder, backbone weights, task head, zero-shot and few-shot learning, linear probing, fine-tuning, LoRa*

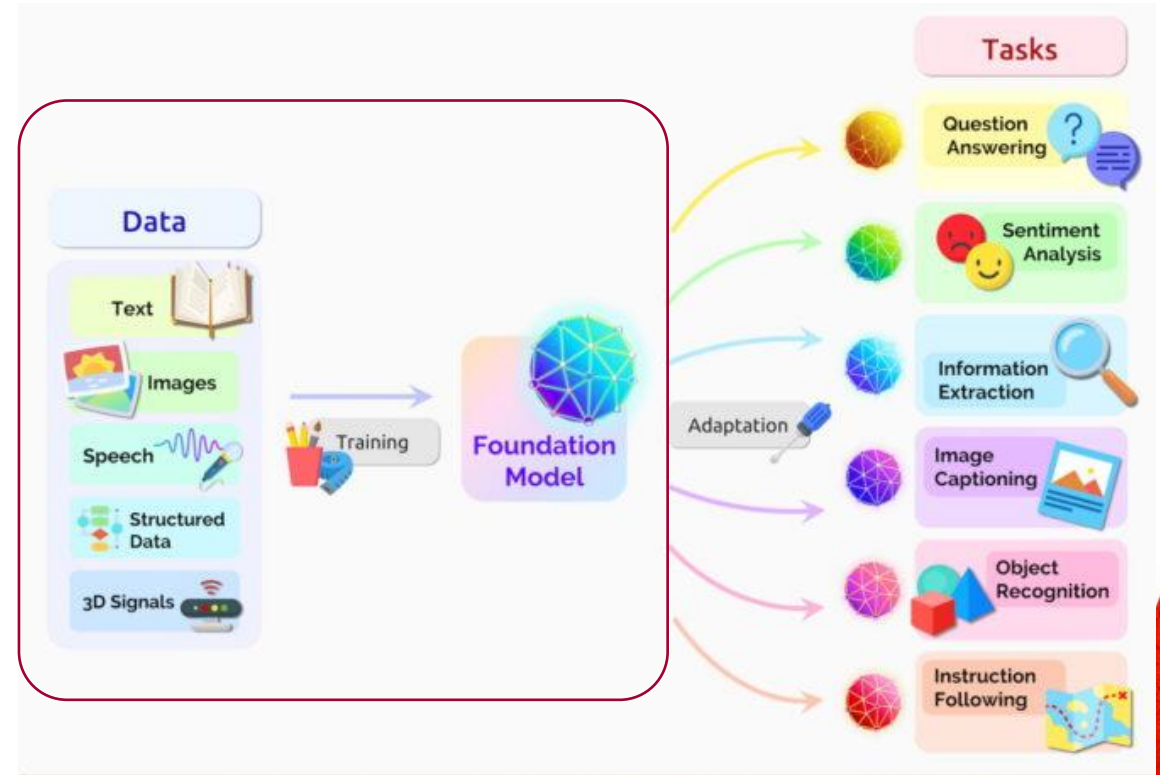
2. Opportunities

- Q&A (if needed!)

“Building” & “Using” BCI Foundation Models

How are Foundation Models trained?

- Typically, on large datasets using self-supervised learning (SSL)
- SSL does not rely on manually labeled data and uses the structure of the data itself to create “learning objectives”.



<https://blogs.nvidia.com/blog/what-are-foundation-models/>

Autoencoders – Preliminaries

- Learn to compress and reconstruct data
- Encoder maps input to a lower dimension (e.g. feature space)
- Decoder reconstructs original input

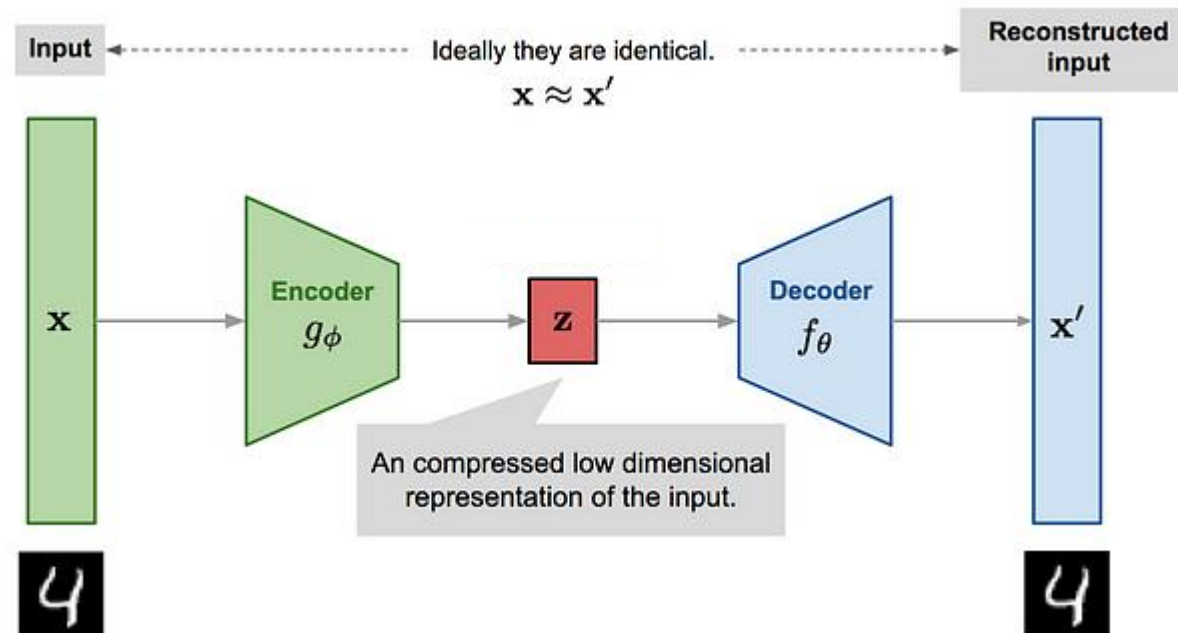
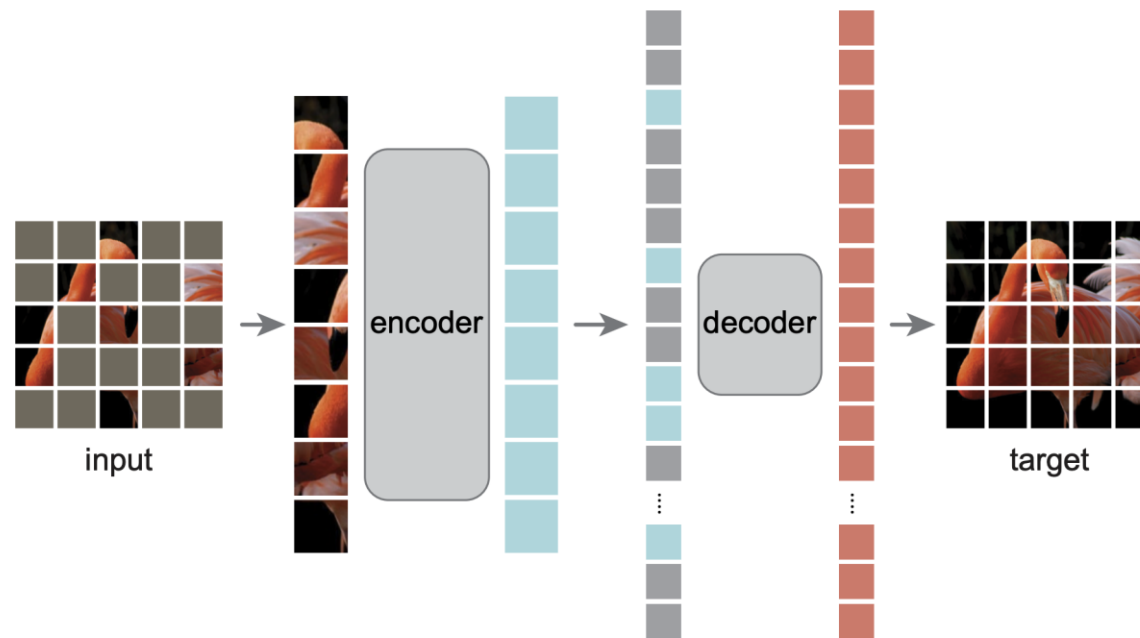


Illustration of the Autoencoder architecture. Source: [Lilianweng's blog](#)

Self-Supervised Learning – *Pretraining without labels*

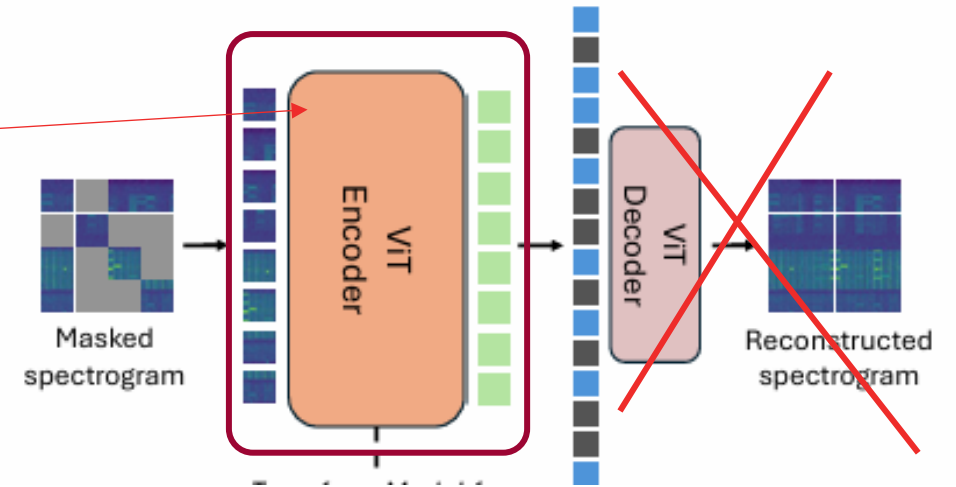
- Form of deep learning that does not require labeled data
- Learns from the data itself using a “pretext task” to generate supervision
 - **Predicting masked portions of input** is a common pretext task



Divide an image into patches and reconstruct the missing patches

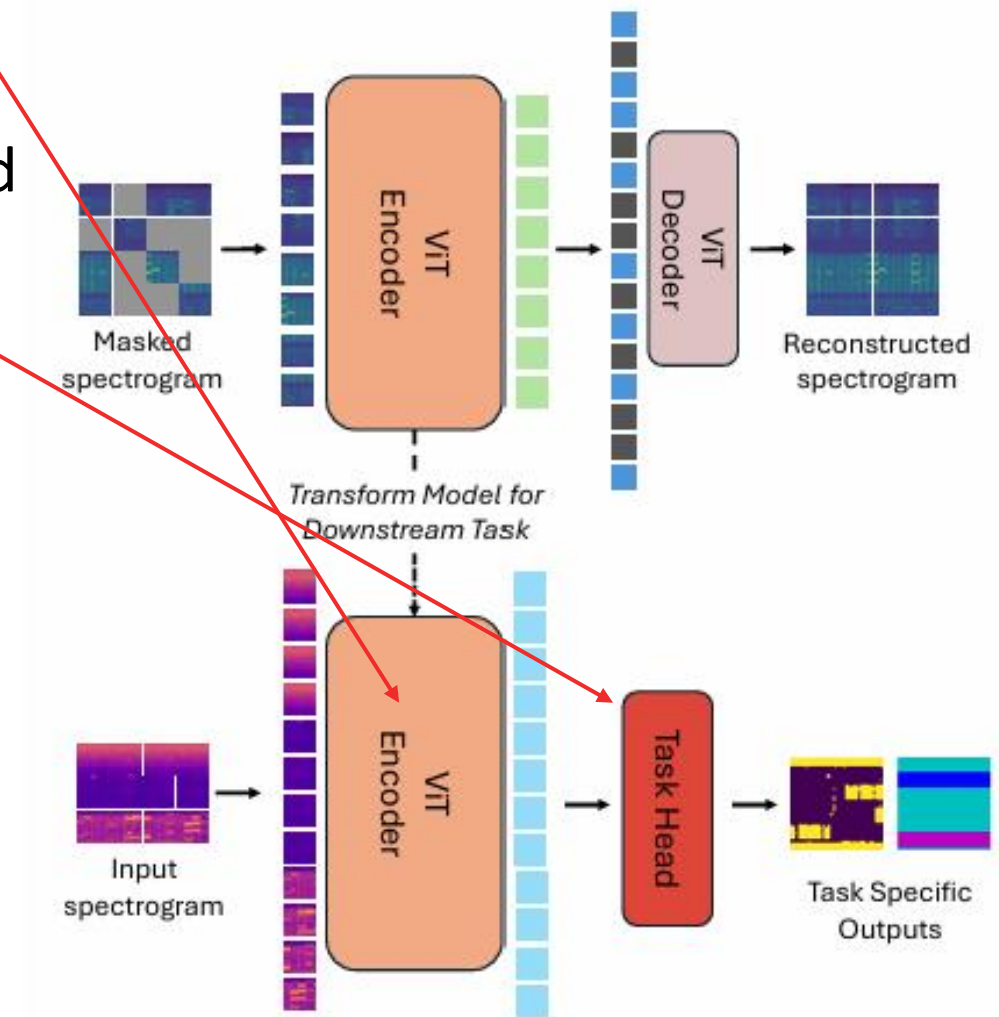
The encoder is the foundation model

- An example for a spectrogram foundation model
- After pretraining, the decoder is discarded
- **The FM is the encoder**



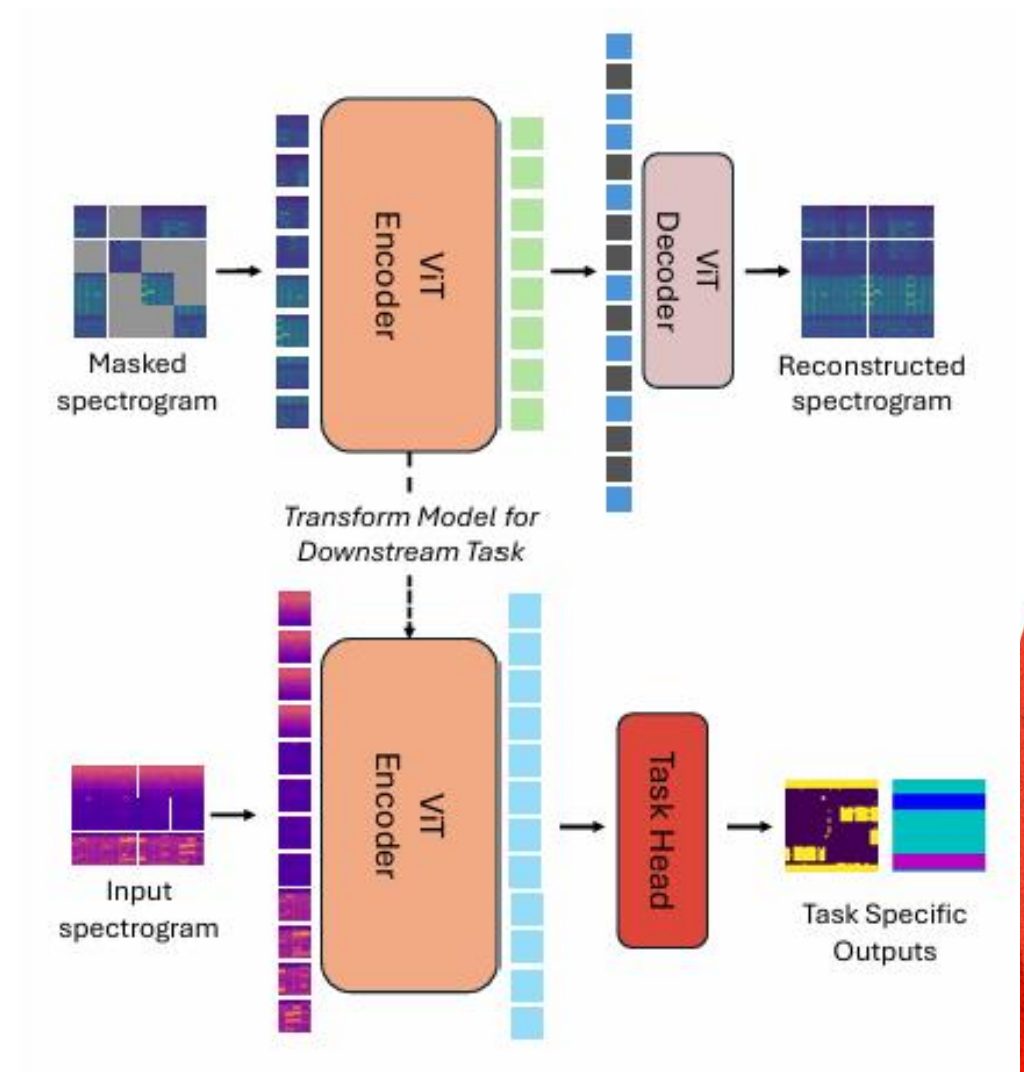
Learning a new task

- Re-use the learned encoder for features
- A simple classifier called the “task head” is added to the model
 - This can be a multi-layer perceptron (MLP) neural network



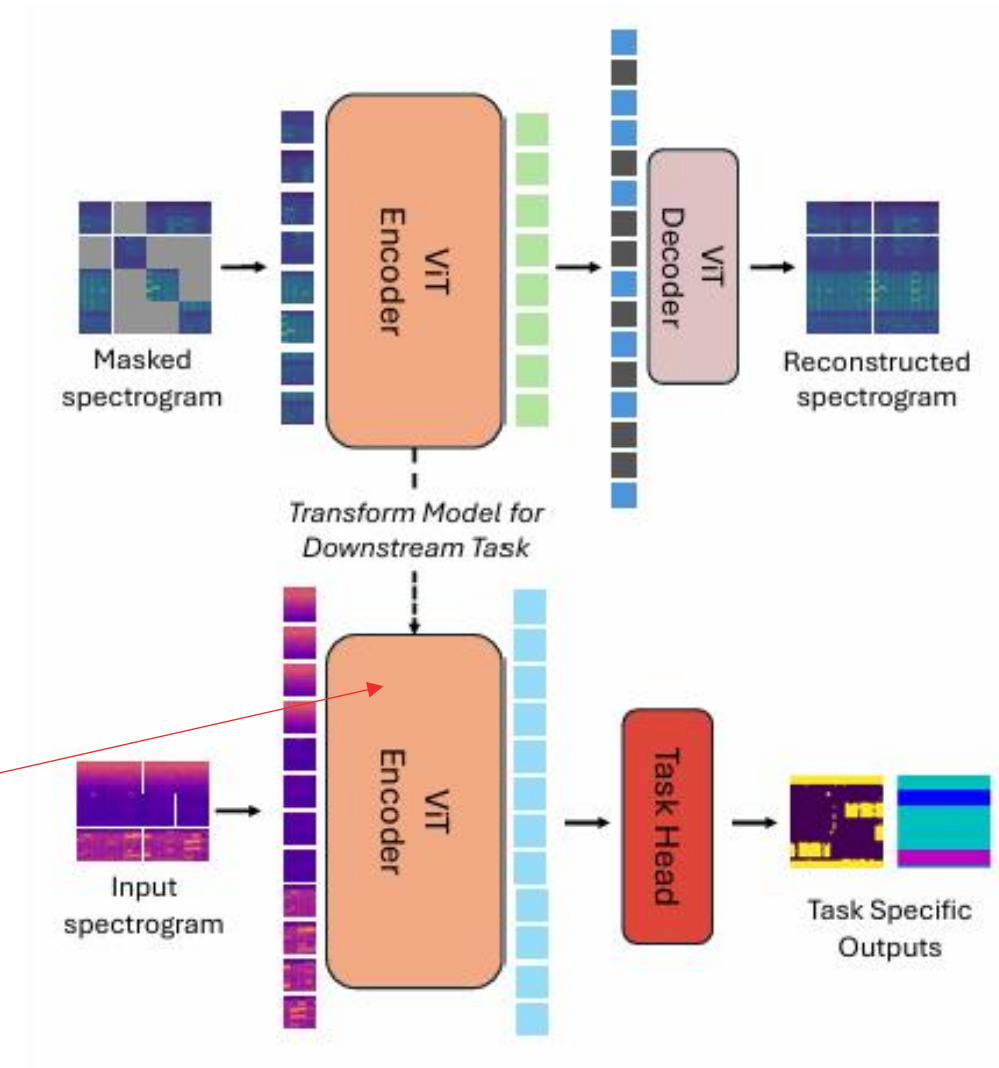
Learning a new task

- The “task head” weights are learned by processing labeled “task” data through the encoder and the task head/classifier
- Note that the encoder is not re-trained here
 - This is commonly termed as a “frozen encoder/FM weights”
- An example is teaching the model to segment spectrograms or classify objects



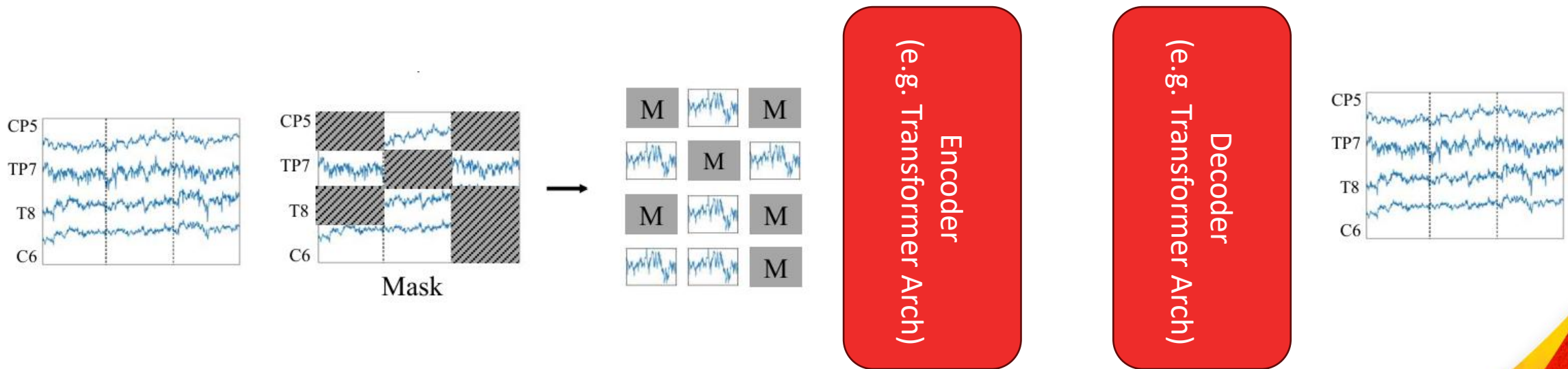
Can we fine-tune the encoder?

- Fine-tuning the encoder is also possible and involves updating some of the encoder weights
 - There are very efficient ways to do this
- When might we need this?



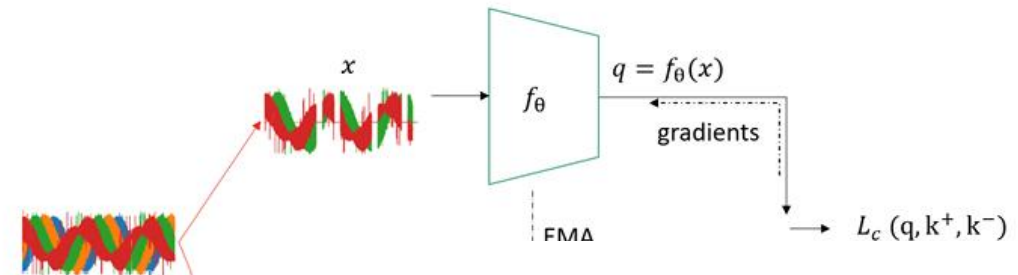
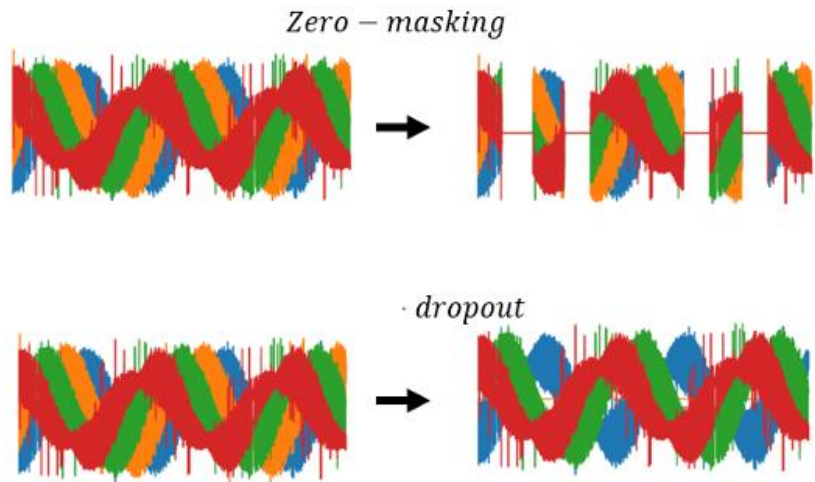
How can we apply this to EEG signals?

- Transform the channel-time EEG signals into “patches” of channels and temporal segments
- Apply a similar Encoder- Decoder structure to reconstruct the masked EEG signals
- Discard the decoder and use the encoder as the foundation model for new BCI tasks

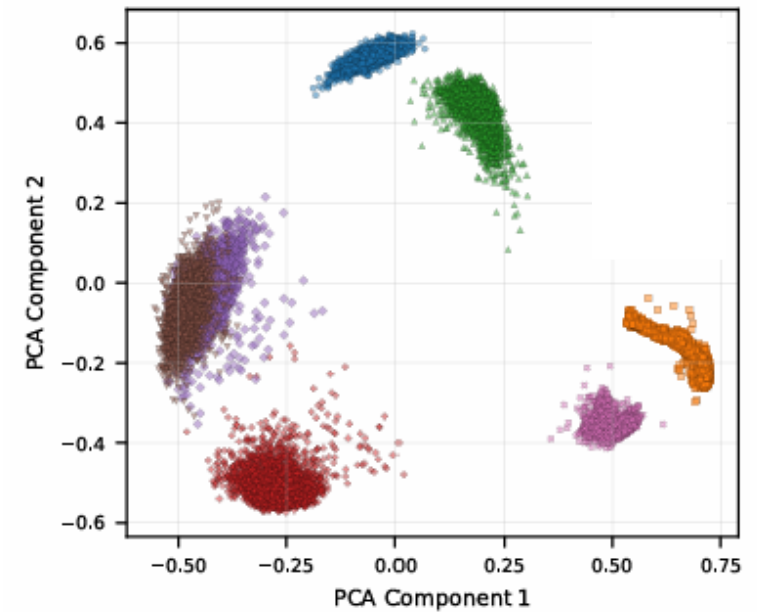


This is one approach but there are many other self-supervised learning approaches and variants!

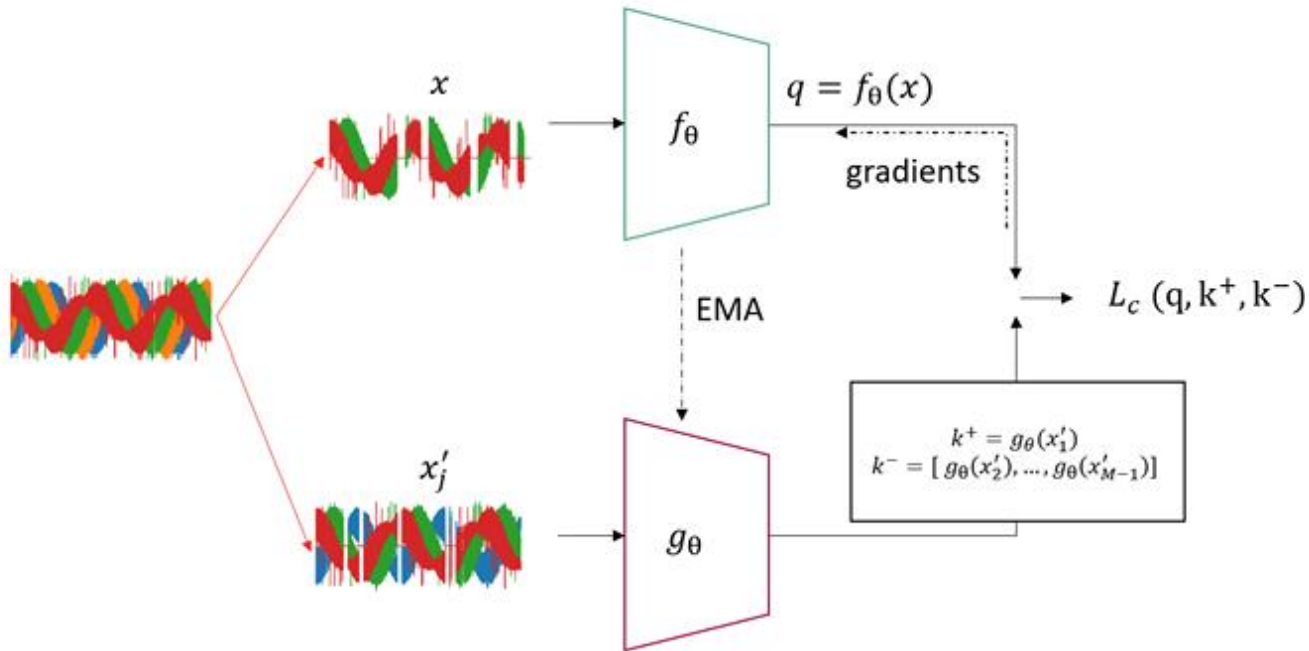
Contrastive Learning – *Signal Augmentations*



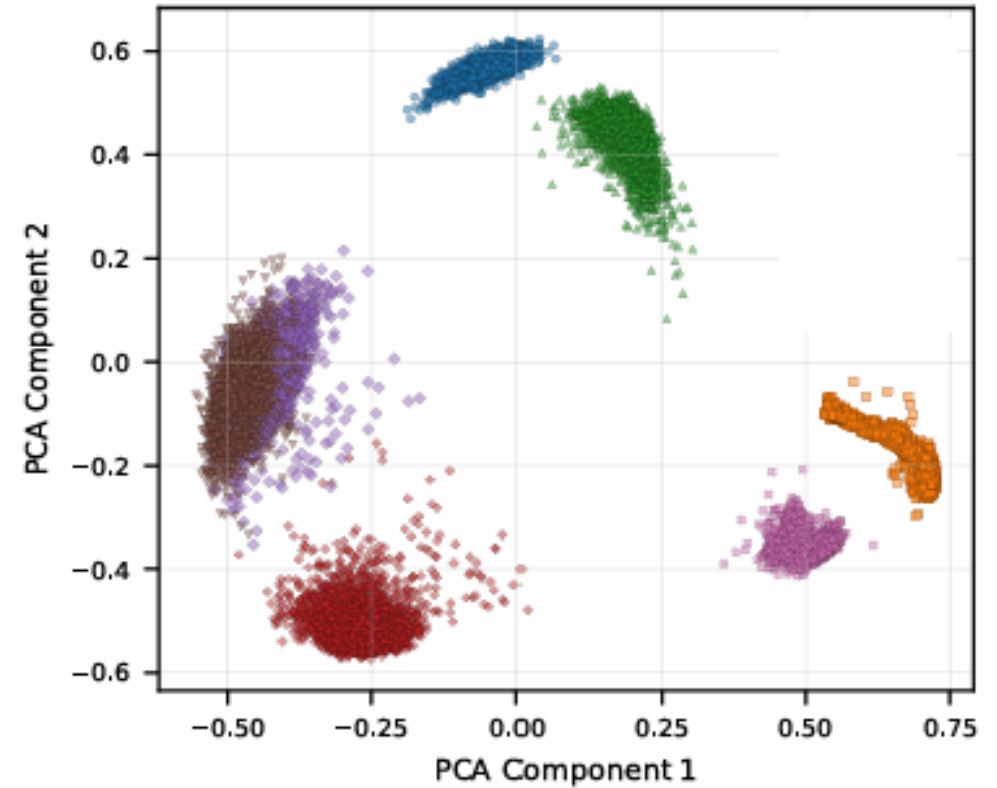
(a) Self-supervis



Contrastive Learning – *Clusters created*



(a) Self-supervised pretraining with MoCo framework.



Opportunities

Thank You!

Questions?

hatem.abouzeid@ucalgary.ca



**UNIVERSITY OF
CALGARY**